

Title: First-Passage Processes in Physics and Beyond

Speakers: Sidney Redner

Series: Colloquium

Date: November 30, 2022 - 2:00 PM

URL: <https://pirsa.org/22110115>

Abstract: A fundamental aspect of a random walk is determining when it reaches a specified threshold position for the first time. This first-passage time, and more generally, the distribution of first passage times underlies many non-equilibrium phenomena, such as the triggering of integrate and fire neurons, the statistics of cell division, and the execution of stock options. The computation of the first-passage time and its distribution is both simple and beautiful, with profound connections to electrostatic potential theory. I will present some aspects of these fundamentals and then discuss applications of first-passage ideas to diverse phenomena, including stochastic search processes and a toy model of wealth sharing.

Zoom link: <https://pitp.zoom.us/j/98293478936?pwd=NTR3dWZoNElWRmd2NVJ1bzk5aC9ZQT09>

# First-Passage Processes in Physics and Beyond

Perimeter Institute, Nov 30, 2022

collaborators: D. ben Avraham, P. L. Krapivsky, B. De Bruyne, J. Random-Furling &  support  
*Sid Redner, Santa Fe Institute, [santafe.edu/~redner](http://santafe.edu/~redner)*

# First-Passage Processes in Physics and Beyond

Perimeter Institute, Nov 30, 2022

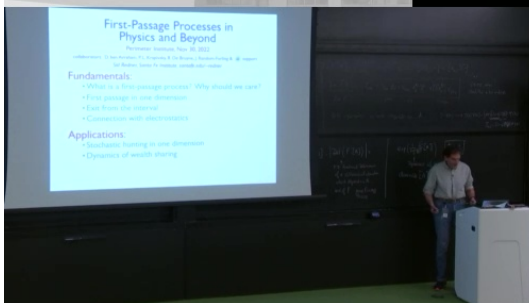
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Sid Redner, Santa Fe Institute, [santafe.edu/~redner](http://santafe.edu/~redner)

## Fundamentals:

- What is a first-passage process? Why should we care?
- First passage in one dimension
- Exit from the interval
- Connection with electrostatics

## Applications:

- Stochastic hunting in one dimension
- Dynamics of wealth sharing

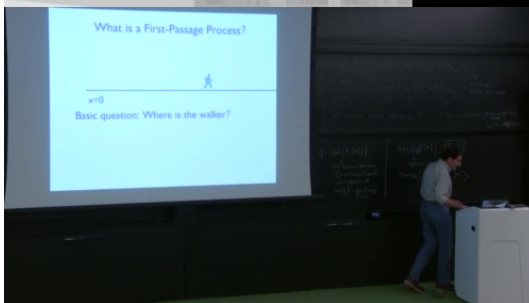


# What is a First-Passage Process?



$x=0$

Basic question: Where is the walker?



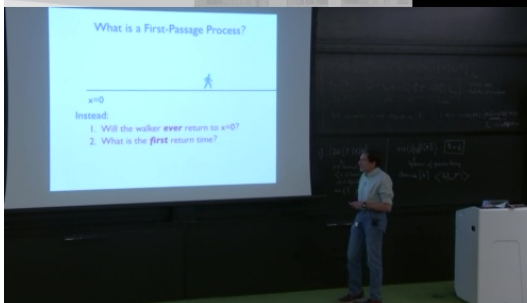
# What is a First-Passage Process?



$x=0$

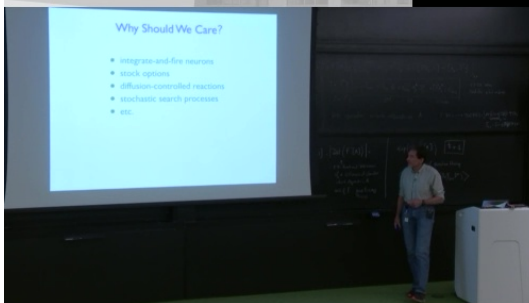
Instead:

1. Will the walker **ever** return to  $x=0$ ?
2. What is the **first** return time?



## Why Should We Care?

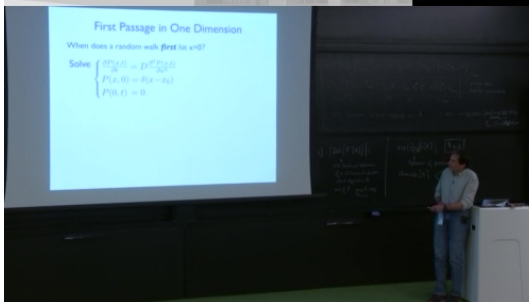
- integrate-and-fire neurons
- stock options
- diffusion-controlled reactions
- stochastic search processes
- etc.



# First Passage in One Dimension

When does a random walk **first** hit  $x=0$ ?

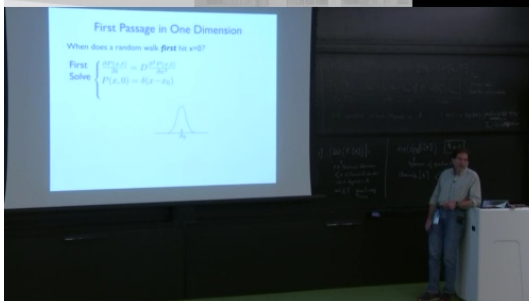
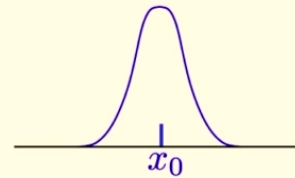
$$\text{Solve } \begin{cases} \frac{\partial P(x,t)}{\partial t} = D \frac{\partial^2 P(x,t)}{\partial x^2} \\ P(x,0) = \delta(x-x_0) \\ P(0,t) = 0 \end{cases}$$



# First Passage in One Dimension

When does a random walk **first** hit  $x=0$ ?

First Solve  $\left\{ \begin{array}{l} \frac{\partial P(x,t)}{\partial t} = D \frac{\partial^2 P(x,t)}{\partial x^2} \\ P(x,0) = \delta(x-x_0) \end{array} \right.$

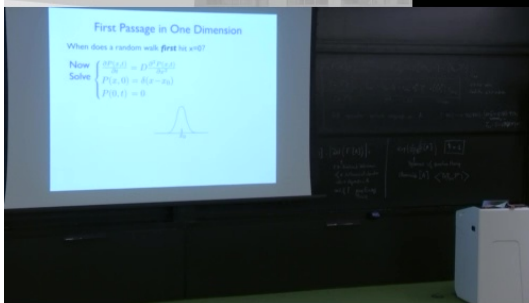
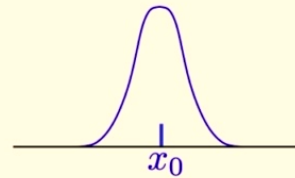




# First Passage in One Dimension

When does a random walk **first** hit  $x=0$ ?

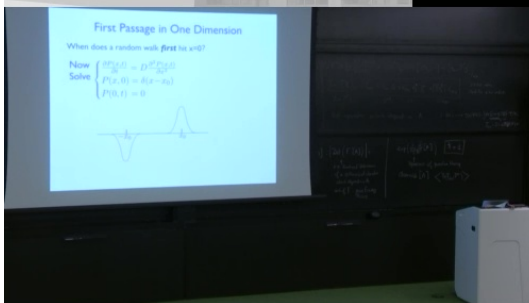
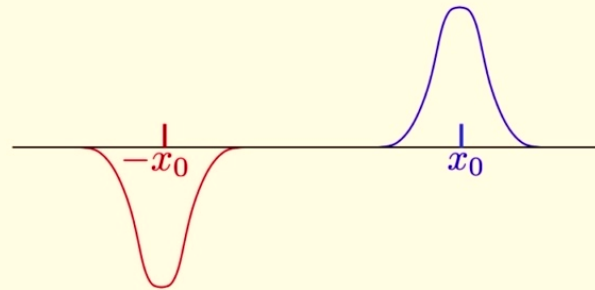
$$\text{Now Solve } \begin{cases} \frac{\partial P(x,t)}{\partial t} = D \frac{\partial^2 P(x,t)}{\partial x^2} \\ P(x,0) = \delta(x-x_0) \\ P(0,t) = 0 \end{cases}$$



# First Passage in One Dimension

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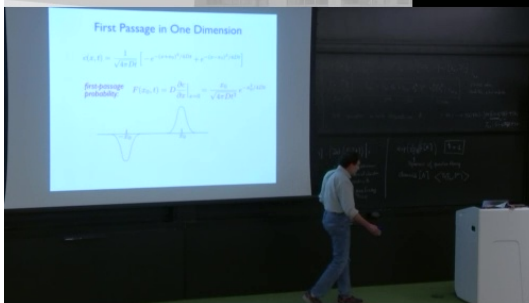
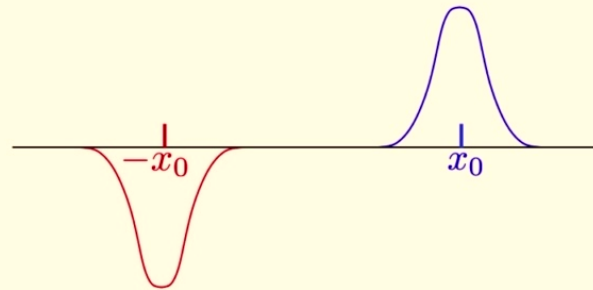
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# First Passage in One Dimension

$$c(x, t) = \frac{1}{\sqrt{4\pi Dt}} \left[ -e^{-(x+x_0)^2/4Dt} + e^{-(x-x_0)^2/4Dt} \right]$$

*first-passage probability:*  $F(x_0, t) = D \frac{\partial c}{\partial x} \Big|_{x=0} = \frac{x_0}{\sqrt{4\pi Dt^3}} e^{-x_0^2/4Dt}$

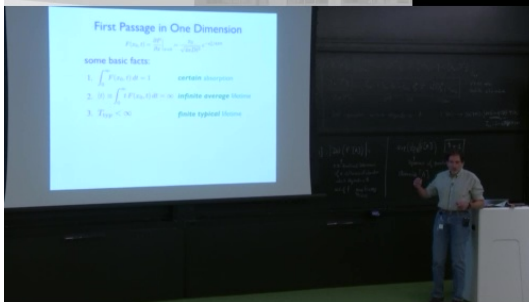


# First Passage in One Dimension

$$F(x_0, t) = \left. \frac{\partial P}{\partial x} \right|_{x=0} = \frac{x_0}{\sqrt{4\pi Dt^3}} e^{-x_0^2/4Dt}$$

some basic facts:

1.  $\int_0^\infty F(x_0, t) dt = 1$  **certain** absorption
2.  $\langle t \rangle \equiv \int_0^\infty t F(x_0, t) dt = \infty$  **infinite average** lifetime
3.  $T_{\text{typ}} < \infty$  **finite typical** lifetime



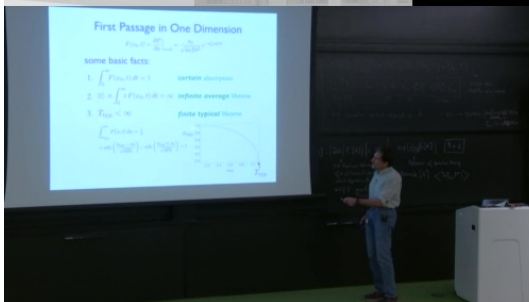
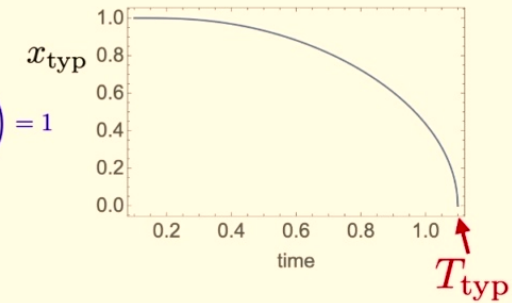
# First Passage in One Dimension

$$F(x_0, t) = \frac{\partial P}{\partial x} \Big|_{x=0} = \frac{x_0}{\sqrt{4\pi Dt^3}} e^{-x_0^2/4Dt}$$

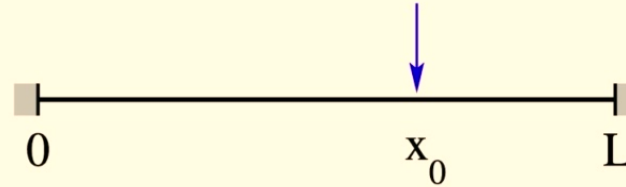
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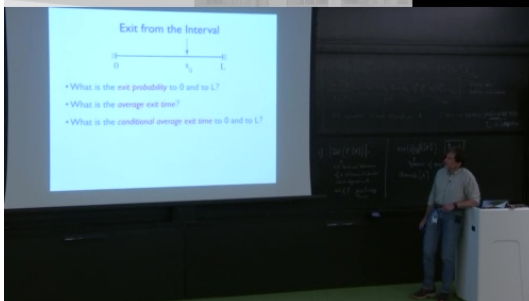
$$\int_{x_{\text{typ}}}^\infty P(x, t) dx = \frac{1}{2}$$
$$\rightarrow \operatorname{erfc}\left(\frac{x_{\text{typ}} - x_0}{\sqrt{4Dt}}\right) - \operatorname{erfc}\left(\frac{x_{\text{typ}} + x_0}{\sqrt{4Dt}}\right) = 1$$



## Exit from the Interval



- What is the *exit probability* to 0 and to L?
- What is the *average exit time*?
- What is the *conditional average exit time* to 0 and to L?

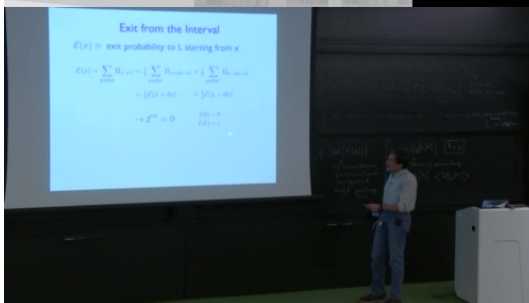


## Exit from the Interval

$\mathcal{E}(x) \equiv$  exit probability to L starting from x

$$\begin{aligned}\mathcal{E}(x) &= \sum_{\text{paths}} \Pi_{x \rightarrow L} = \frac{1}{2} \sum_{\text{paths}' } \Pi_{x+dx \rightarrow L} + \frac{1}{2} \sum_{\text{paths}''} \Pi_{x-dx \rightarrow L} \\ &= \frac{1}{2} \mathcal{E}(x+dx) + \frac{1}{2} \mathcal{E}(x-dx)\end{aligned}$$

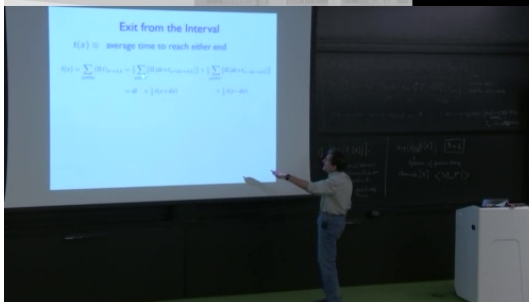
$$\begin{aligned}\rightarrow \mathcal{E}'' &= 0 & \mathcal{E}(0) &= 0 \\ & & \mathcal{E}(L) &= 1\end{aligned}$$



## Exit from the Interval

$t(x) \equiv$  average time to reach *either* end

$$\begin{aligned} t(x) &= \sum_{\text{paths}} (\Pi t)_{x \rightarrow \pm L} = \frac{1}{2} \sum_{\text{paths}' } \{ \Pi [dt + t_{x+dx \rightarrow \pm L}] \} + \frac{1}{2} \sum_{\text{paths}'' } \{ \Pi [dt + t_{x-dx \rightarrow \pm L}] \} \\ &= dt + \frac{1}{2} t(x+dx) + \frac{1}{2} t(x-dx) \end{aligned}$$





## Exit from the Interval

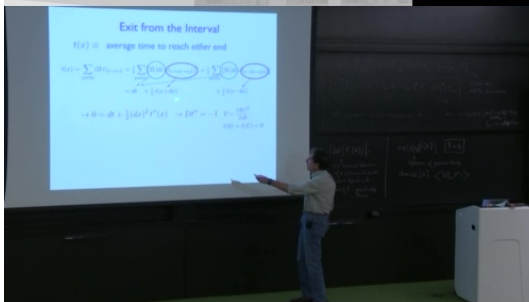
$t(x) \equiv$  average time to reach either end

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$$= dt + \frac{1}{2} t(x+dx) + \frac{1}{2} t(x-dx)$$

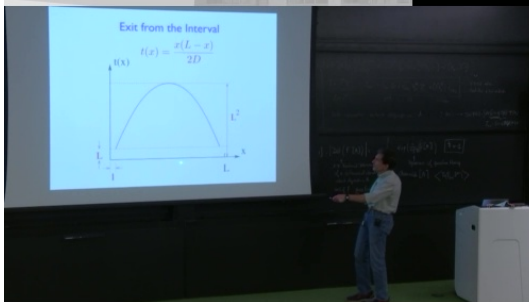
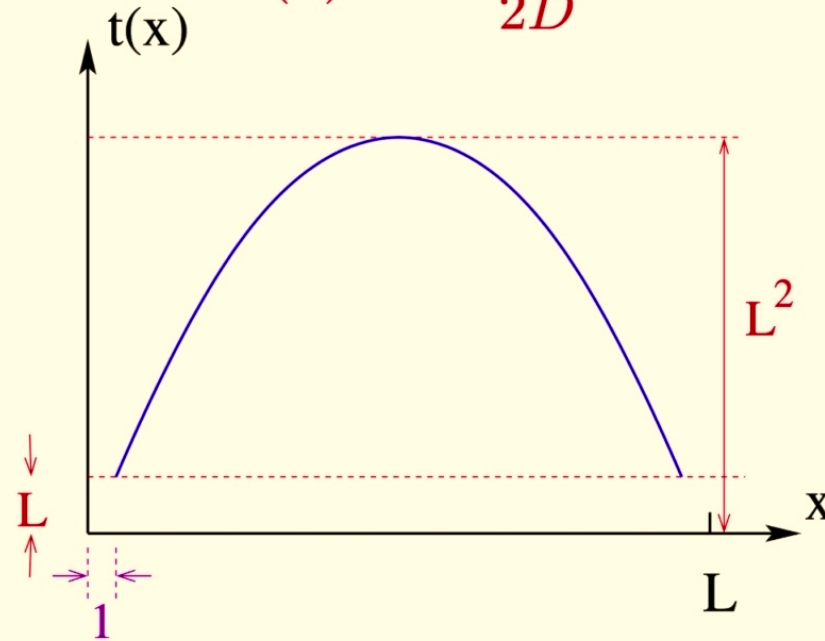
$$\rightarrow 0 = dt + \frac{1}{2} (dx)^2 t''(x) \quad \rightarrow Dt'' = -1 \quad D = \frac{(dx)^2}{2 dt}$$

$$t(0) = t(L) = 0$$



## Exit from the Interval

$$t(x) = \frac{x(L-x)}{2D}$$



## Exit from the Interval

$t(x) \equiv$  average time to reach either end

$$t(x) = \sum_{\text{paths}} (\Pi t)_{x \rightarrow \pm L} = \frac{1}{2} \sum_{\text{paths}'_+} \{ \Pi [dt + t_{x+dx \rightarrow \pm L}] \} + \frac{1}{2} \sum_{\text{paths}'_-} \{ \Pi [dt + t_{x-dx \rightarrow \pm L}] \}$$

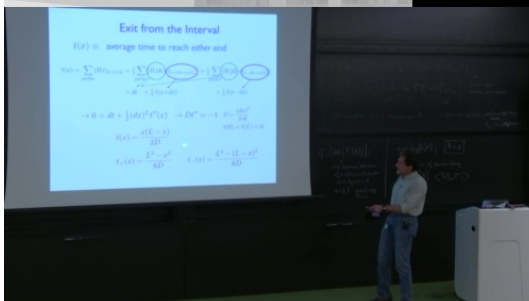
$$= dt + \frac{1}{2} t(x+dx) + \frac{1}{2} t(x-dx)$$

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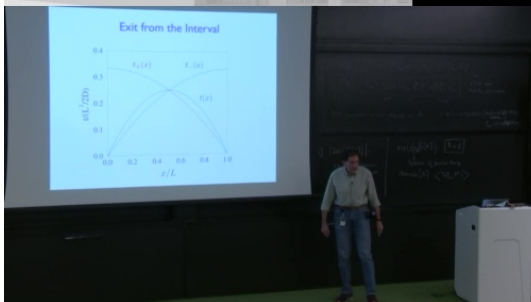
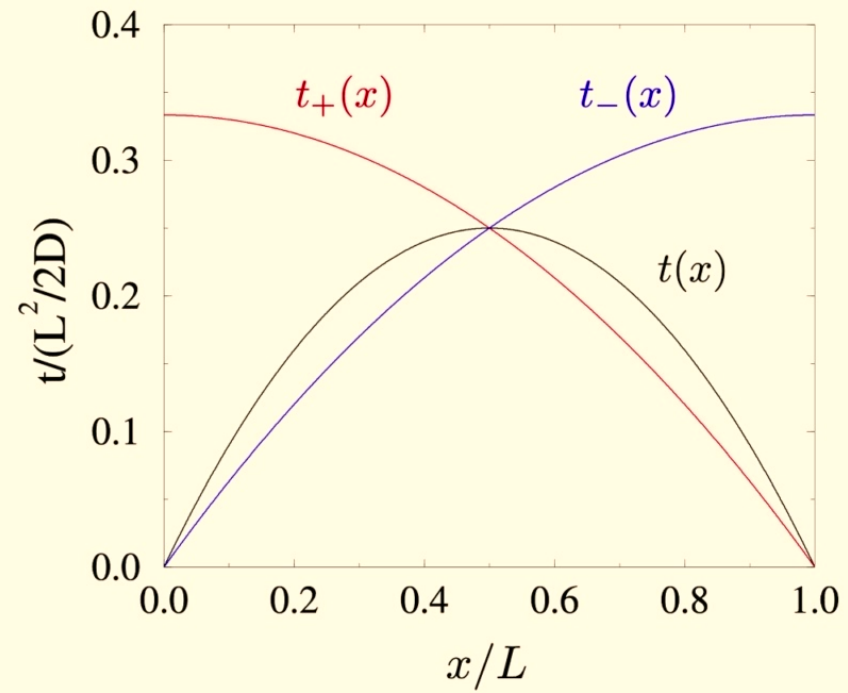
$$t(0) = t(L) = 0$$

$$t(x) = \frac{x(L-x)}{2D}$$

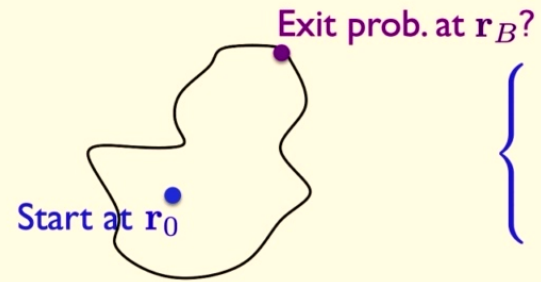
$$t_+(x) = \frac{L^2 - x^2}{6D} \quad t_-(x) = \frac{L^2 - (L-x)^2}{6D}$$



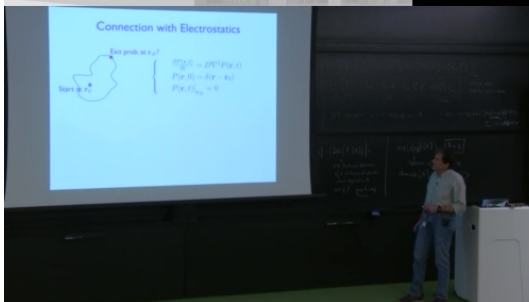
## Exit from the Interval



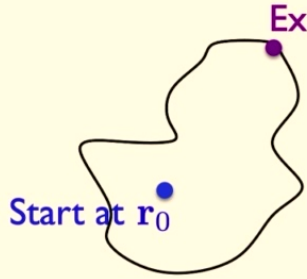
## Connection with Electrostatics



$$\left\{ \begin{array}{l} \frac{\partial P(\mathbf{r}, t)}{\partial t} = D \nabla^2 P(\mathbf{r}, t) \\ P(\mathbf{r}, 0) = \delta(\mathbf{r} - \mathbf{r}_0) \\ P(\mathbf{r}, t)|_{\mathbf{r}_B} = 0 \end{array} \right.$$



## Connection with Electrostatics

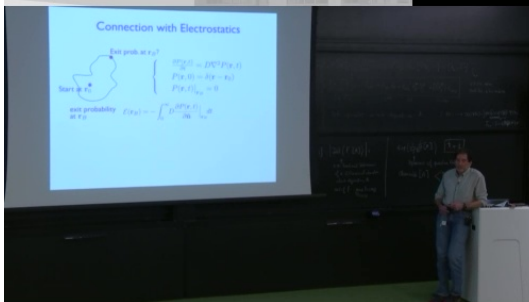


Start at  $\mathbf{r}_0$

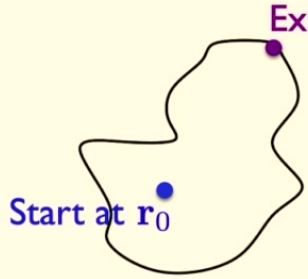
Exit prob. at  $\mathbf{r}_B$ ?

$$\left\{ \begin{array}{l} \frac{\partial P(\mathbf{r}, t)}{\partial t} = D \nabla^2 P(\mathbf{r}, t) \\ P(\mathbf{r}, 0) = \delta(\mathbf{r} - \mathbf{r}_0) \\ P(\mathbf{r}, t)|_{\mathbf{r}_B} = 0 \end{array} \right.$$

exit probability at  $\mathbf{r}_B$   $\mathcal{E}(\mathbf{r}_B) = - \int_0^\infty D \frac{\partial P(\mathbf{r}, t)}{\partial \hat{\mathbf{n}}} \Big|_{\mathbf{r}_B} dt$



## Connection with Electrostatics



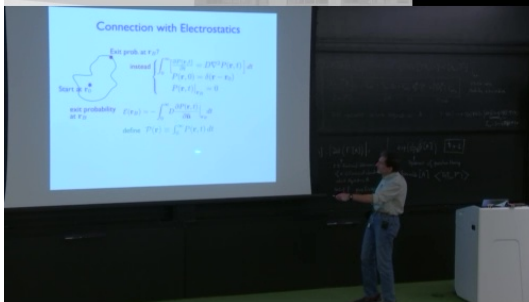
Exit prob. at  $\mathbf{r}_B$ ?

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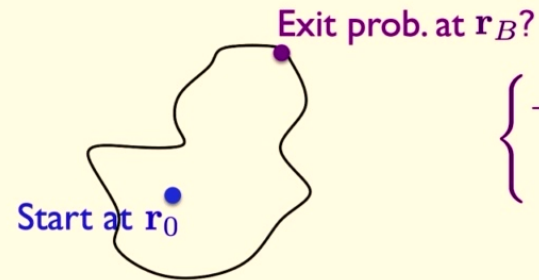
instead  $\left\{ \begin{array}{l} \int_0^\infty \left[ \frac{\partial P(\mathbf{r}, t)}{\partial t} = D \nabla^2 P(\mathbf{r}, t) \right] dt \\ P(\mathbf{r}, 0) = \delta(\mathbf{r} - \mathbf{r}_0) \\ P(\mathbf{r}, t)|_{\mathbf{r}_B} = 0 \end{array} \right.$

exit probability at  $\mathbf{r}_B$   $\mathcal{E}(\mathbf{r}_B) = - \int_0^\infty D \frac{\partial P(\mathbf{r}, t)}{\partial \hat{\mathbf{n}}} \Big|_{\mathbf{r}_B} dt$

define  $\mathcal{P}(\mathbf{r}) \equiv \int_0^\infty P(\mathbf{r}, t) dt$



## Connection with Electrostatics

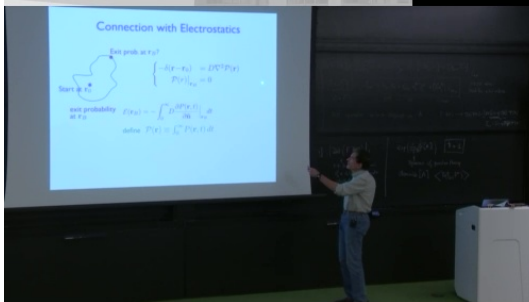


$$\begin{cases} -\delta(\mathbf{r}-\mathbf{r}_0) = D\nabla^2\mathcal{P}(\mathbf{r}) \\ \mathcal{P}(\mathbf{r})|_{\mathbf{r}_B} = 0 \end{cases}$$

exit probability  
at  $\mathbf{r}_B$

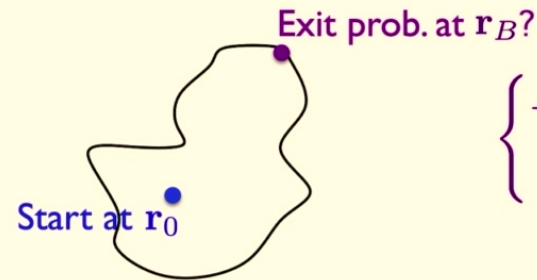
$$\mathcal{E}(\mathbf{r}_B) = -\int_0^\infty D \frac{\partial P(\mathbf{r}, t)}{\partial \hat{\mathbf{n}}} \Big|_{\mathbf{r}_B} dt$$

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## Connection with Electrostatics



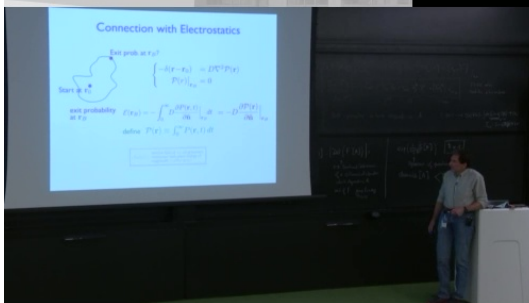
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exit probability at  $\mathbf{r}_B$

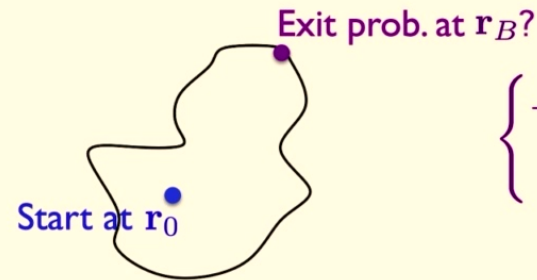
$$\mathcal{E}(\mathbf{r}_B) = -\int_0^\infty D \frac{\partial P(\mathbf{r}, t)}{\partial \hat{\mathbf{n}}} \Big|_{\mathbf{r}_B} dt = -D \frac{\partial \mathcal{P}(\mathbf{r})}{\partial \hat{\mathbf{n}}} \Big|_{\mathbf{r}_B}$$

define  $\mathcal{P}(\mathbf{r}) \equiv \int_0^\infty P(\mathbf{r}, t) dt$

$\mathcal{E}(\mathbf{r}_B) =$  electric field at  $\mathbf{r}_B$  on grounded conductor with point charge of magnitude  $1/(D\Omega_d)$  at  $\mathbf{r}_0$



## Connection with Electrostatics



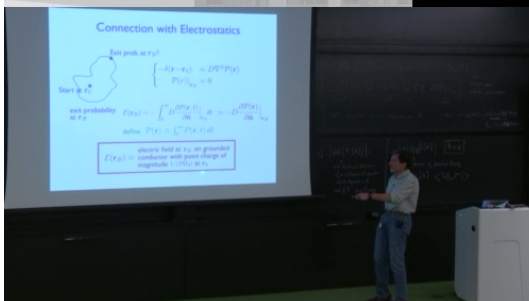
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exit probability at  $\mathbf{r}_B$

$$\mathcal{E}(\mathbf{r}_B) = -\int_0^\infty D \frac{\partial P(\mathbf{r}, t)}{\partial \hat{\mathbf{n}}} \Big|_{\mathbf{r}_B} dt = -D \frac{\partial \mathcal{P}(\mathbf{r})}{\partial \hat{\mathbf{n}}} \Big|_{\mathbf{r}_B}$$

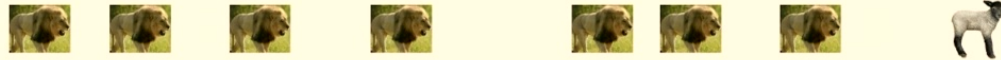
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# Stochastic Hunting in One Dimension

Krapivsky & SR (1999)  
ben-Avraham et al (2003)



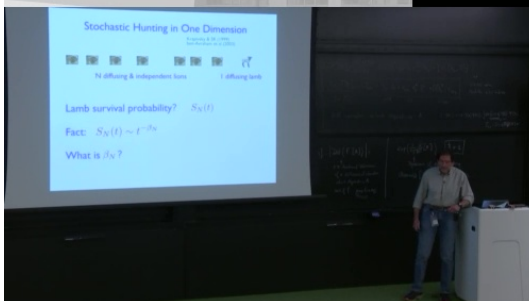
N diffusing & independent lions

1 diffusing lamb

Lamb survival probability?  $S_N(t)$

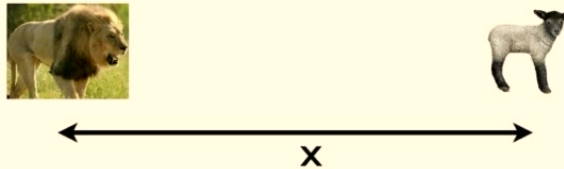
Fact:  $S_N(t) \sim t^{-\beta_N}$

What is  $\beta_N$ ?



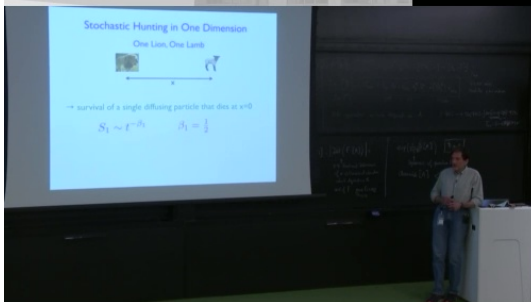
# Stochastic Hunting in One Dimension

One Lion, One Lamb



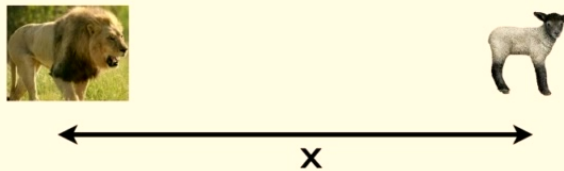
→ survival of a single diffusing particle that dies at  $x=0$

$$S_1 \sim t^{-\beta_1} \quad \beta_1 = \frac{1}{2}$$



# Stochastic Hunting in One Dimension

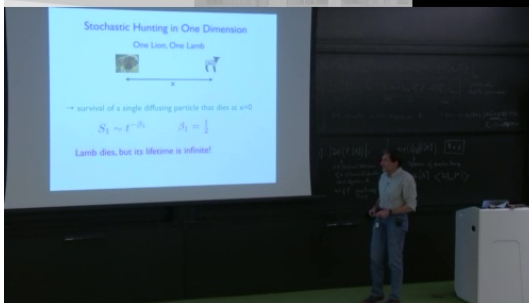
One Lion, One Lamb



→ survival of a single diffusing particle that dies at  $x=0$

$$S_1 \sim t^{-\beta_1} \quad \beta_1 = \frac{1}{2}$$

Lamb dies, but its lifetime is infinite!



# Stochastic Hunting in One Dimension

## Two Lions, One Lamb



$x_1$

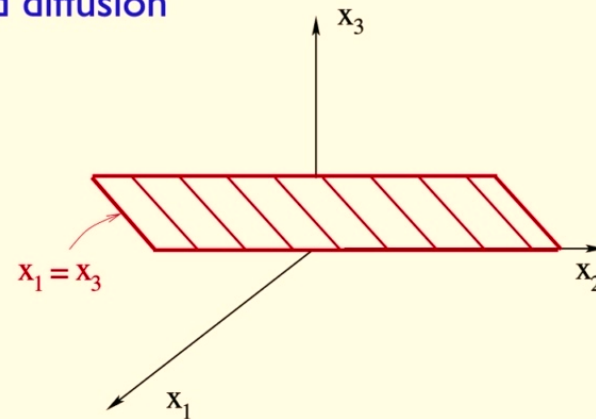


$x_2$



$x_3$

require  $x_1 < x_3$  and  $x_2 < x_3$   
map to 3d diffusion



# Stochastic Hunting in One Dimension

## Two Lions, One Lamb



$x_1$

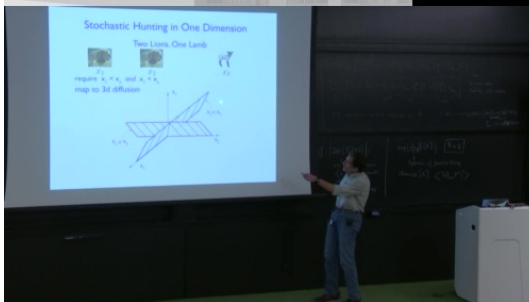
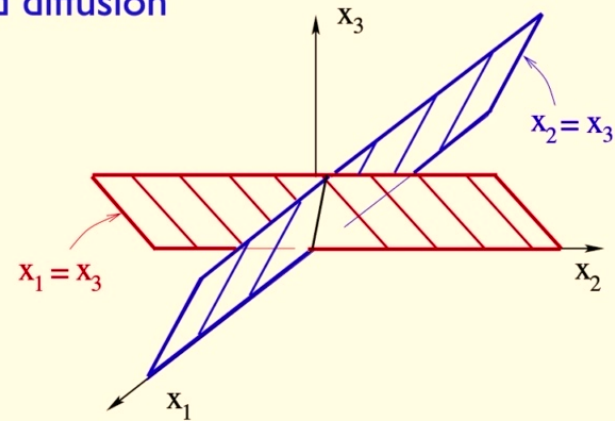


$x_2$



$x_3$

require  $x_1 < x_3$  and  $x_2 < x_3$   
map to 3d diffusion



# Stochastic Hunting in One Dimension

## Two Lions, One Lamb



$x_1$



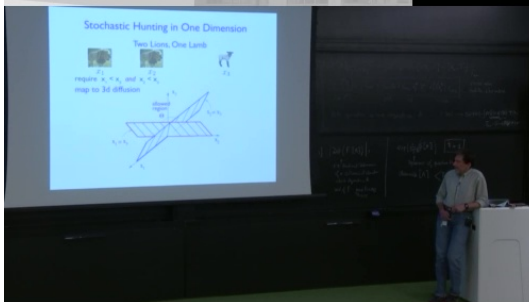
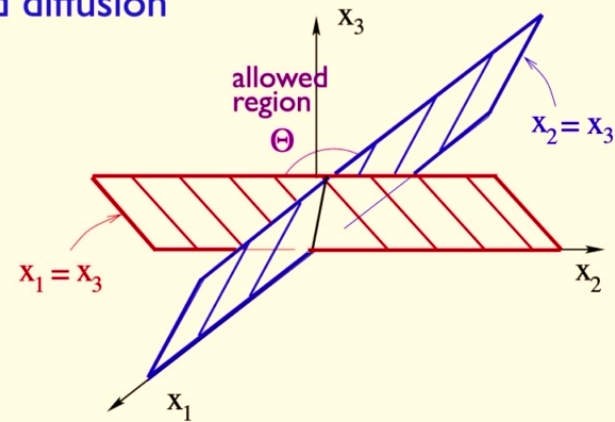
$x_2$



$x_3$

require  $x_1 < x_3$  and  $x_2 < x_3$

map to 3d diffusion





# Stochastic Hunting in One Dimension

## Two Lions, One Lamb



$x_1$

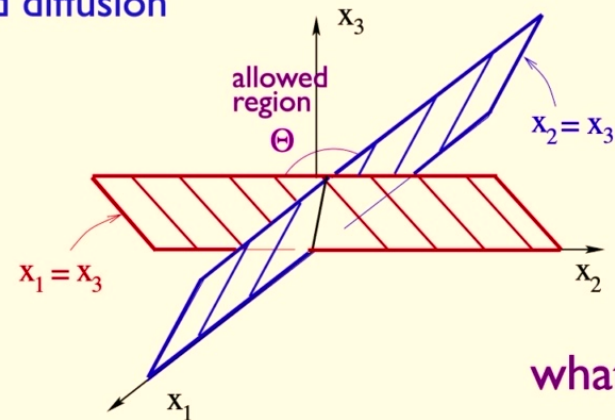


$x_2$

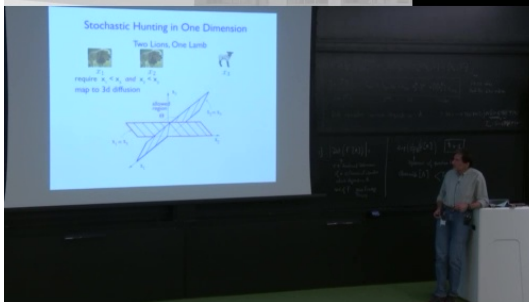


$x_3$

require  $x_1 < x_3$  and  $x_2 < x_3$   
map to 3d diffusion



what is  $\Theta$ ?



# Stochastic Hunting in One Dimension

## Two Lions, One Lamb



$x_1$

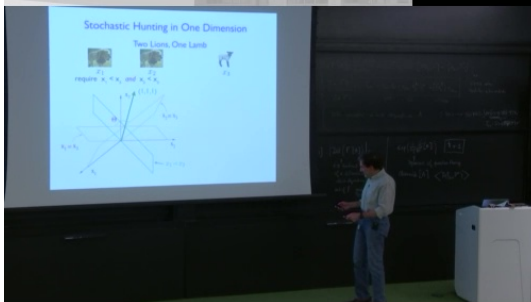
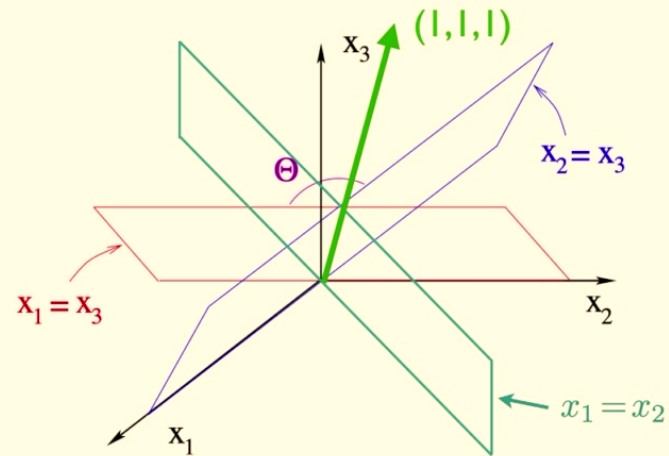


$x_2$



$x_3$

require  $x_1 < x_3$  and  $x_2 < x_3$



# Stochastic Hunting in One Dimension

## Two Lions, One Lamb



$x_1$

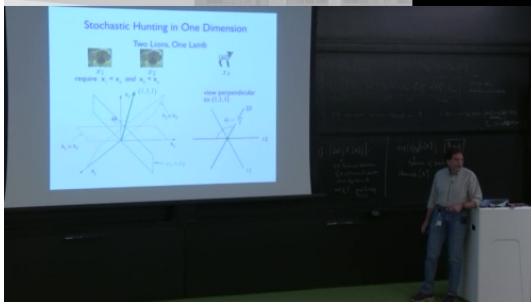
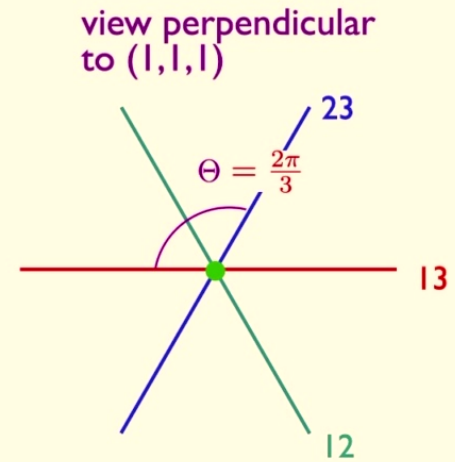
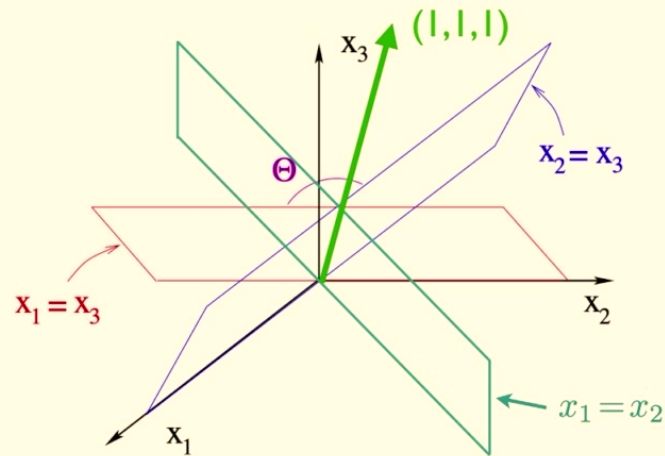


$x_2$



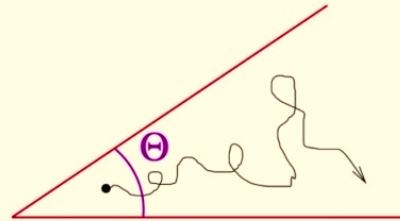
$x_3$

require  $x_1 < x_3$  and  $x_2 < x_3$



# Stochastic Hunting in One Dimension

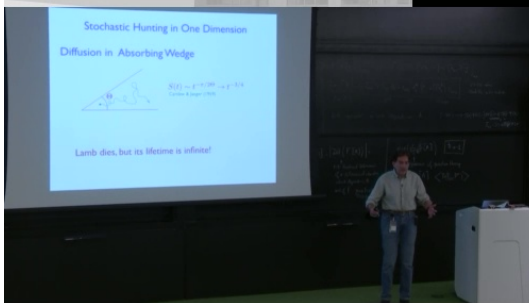
## Diffusion in Absorbing Wedge



$$S(t) \sim t^{-\pi/2\Theta} \rightarrow t^{-3/4}$$

Carslaw & Jaeger (1959)

**Lamb dies, but its lifetime is infinite!**



# Stochastic Hunting in One Dimension

## Three Lions, One Lamb



$$x_1 < x_4$$



$$x_2 < x_4$$



$$x_3 < x_4$$

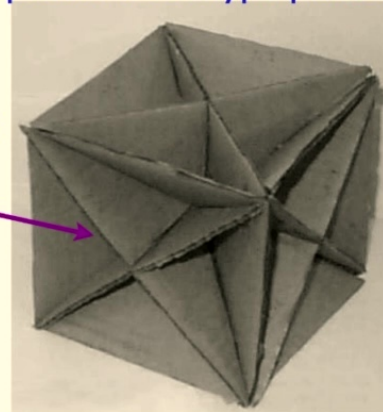


$$x_4$$

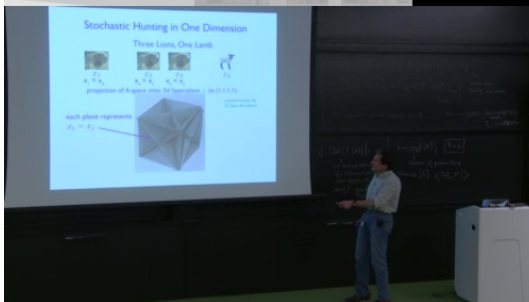
projection of 4-space onto 3d hyperplane  $\perp$  to (1,1,1,1)

each plane represents

$$x_i = x_j$$



construction by  
D. ben-Avraham



# Stochastic Hunting in One Dimension

## Three Lions, One Lamb



$$x_1 < x_4$$



$$x_2 < x_4$$



$$x_3 < x_4$$

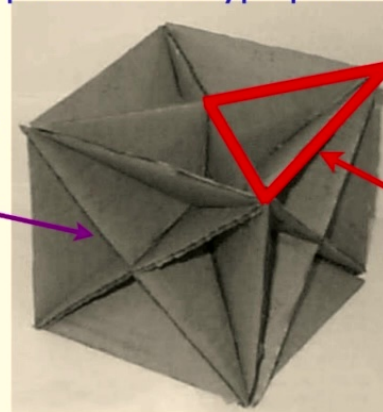


$$x_4$$

projection of 4-space onto 3d hyperplane  $\perp$  to (1,1,1,1)

each plane represents

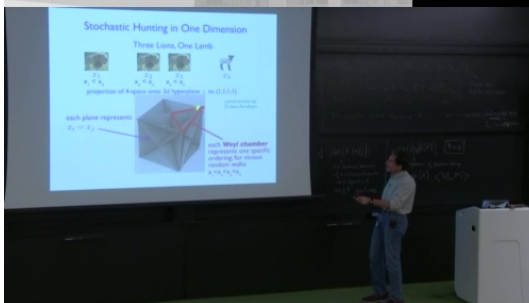
$$x_i = x_j$$



construction by  
D. ben-Avraham

each **Weyl chamber**  
represents one specific  
ordering; for vicious  
random walks

$$x_1 < x_2 < x_3 < x_4$$



# Stochastic Hunting in One Dimension

## Three Lions, One Lamb



$$x_1 < x_4$$



$$x_2 < x_4$$



$$x_3 < x_4$$

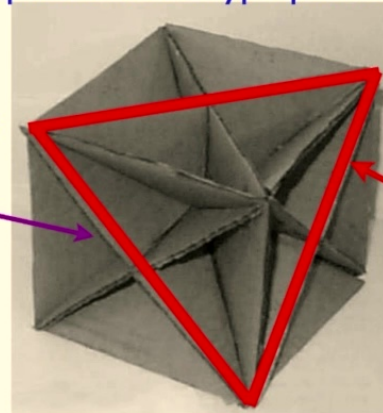


$$x_4$$

projection of 4-space onto 3d hyperplane  $\perp$  to (1,1,1,1)

each plane represents

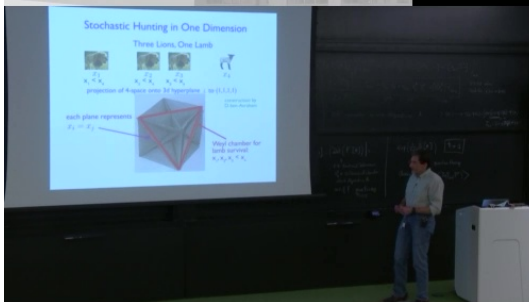
$$x_i = x_j$$



construction by  
D. ben-Avraham

Weyl chamber for  
lamb survival:

$$x_1, x_2, x_3 < x_4$$



# Stochastic Hunting in One Dimension

## Three Lions, One Lamb



$$x_1$$

$$x_1 < x_4$$



$$x_2$$

$$x_2 < x_4$$



$$x_3$$

$$x_3 < x_4$$



$$x_4$$

projection of 4-space onto 3d hyperplane  $\perp$  to (1,1,1,1)

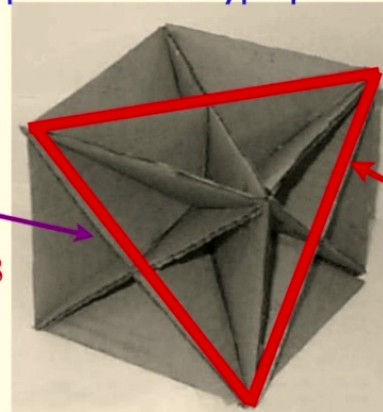
each plane represents

$$x_i = x_j$$

$$\rightarrow \beta_3 = 0.91342 \pm 0.00008$$

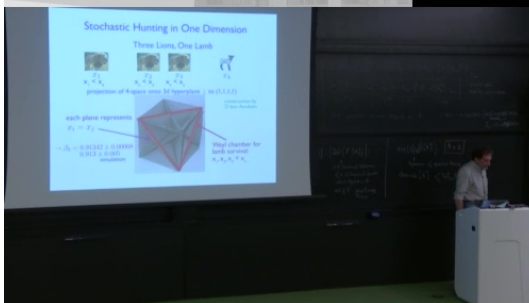
$$0.913 \pm 0.005$$

simulation



construction by  
D. ben-Avraham

Weyl chamber for  
lamb survival:  
 $x_1, x_2, x_3 < x_4$

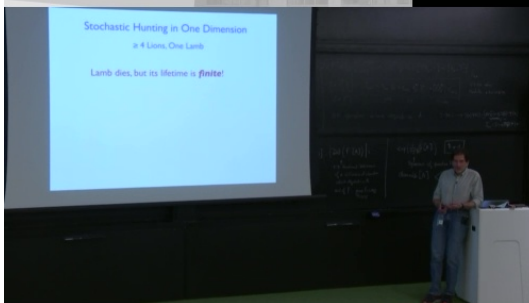




# Stochastic Hunting in One Dimension

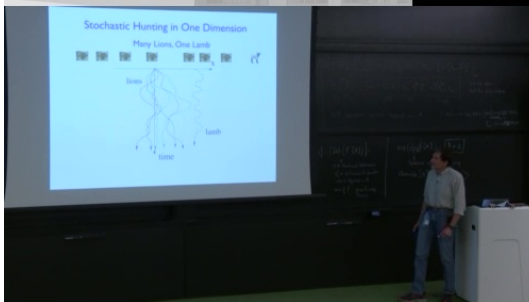
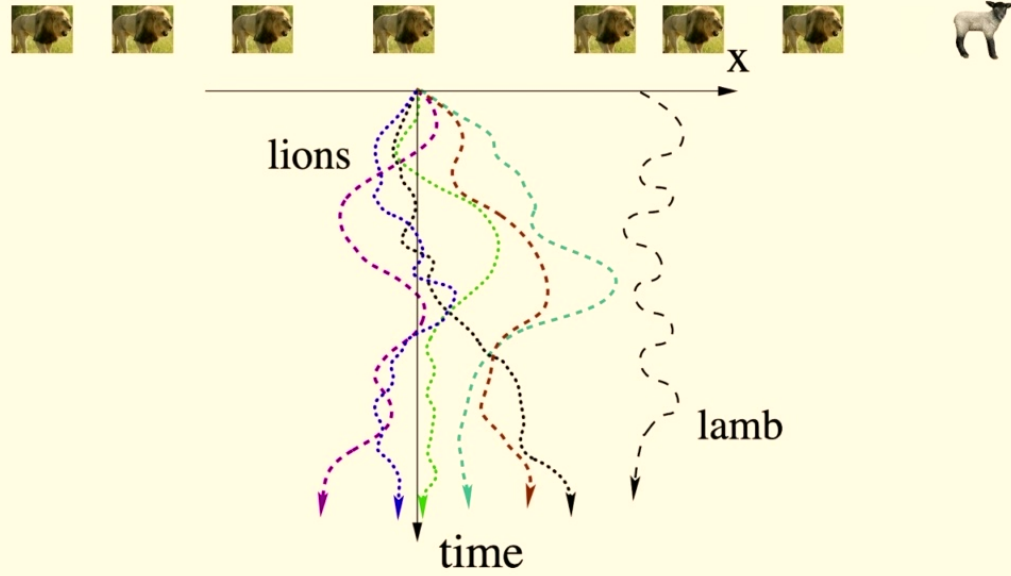
$\geq 4$  Lions, One Lamb

Lamb dies, but its lifetime is ***finite!***



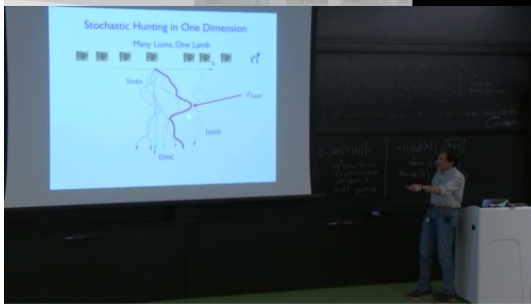
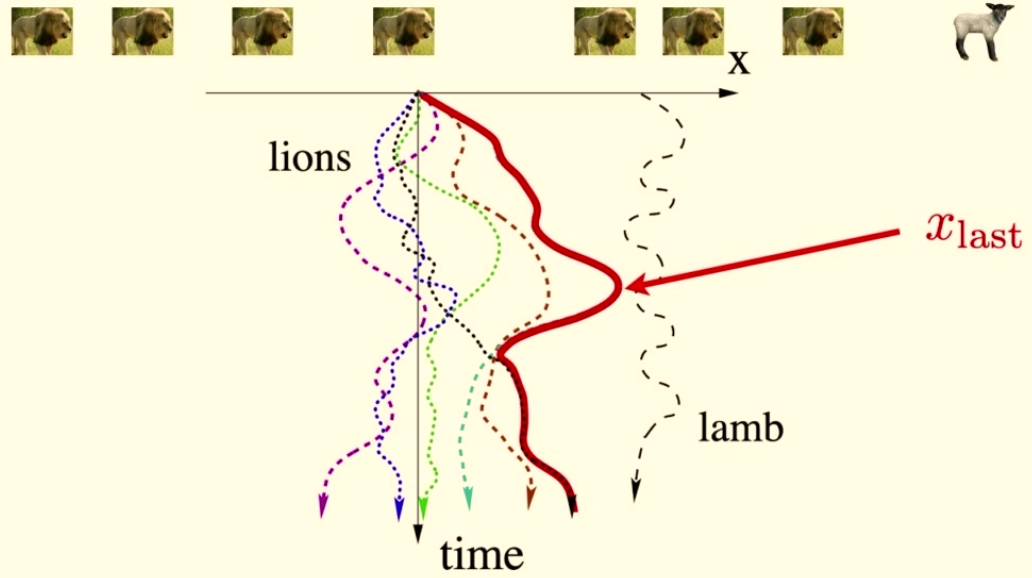
# Stochastic Hunting in One Dimension

Many Lions, One Lamb



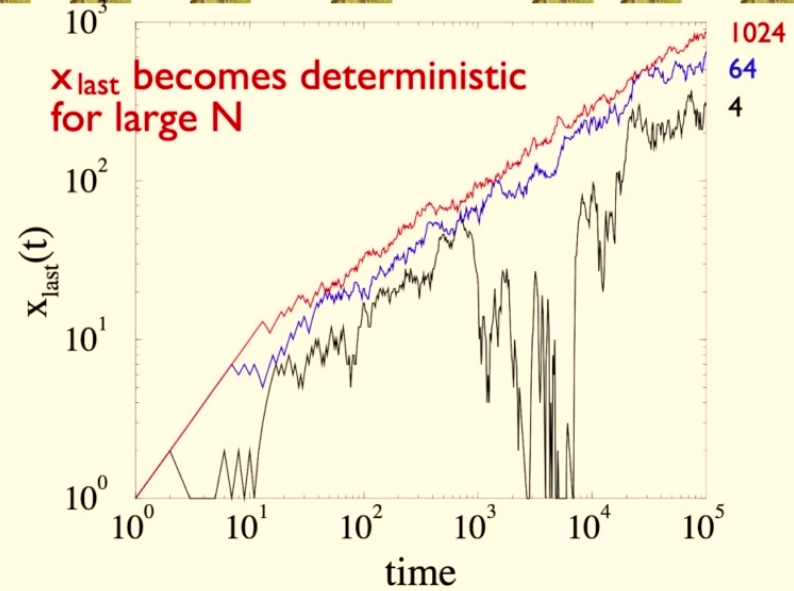
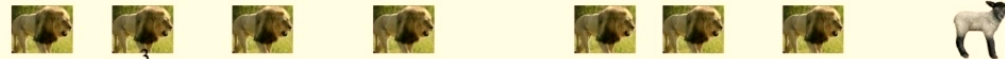
# Stochastic Hunting in One Dimension

Many Lions, One Lamb



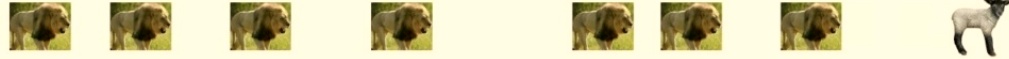
# Stochastic Hunting in One Dimension

Many Lions, One Lamb



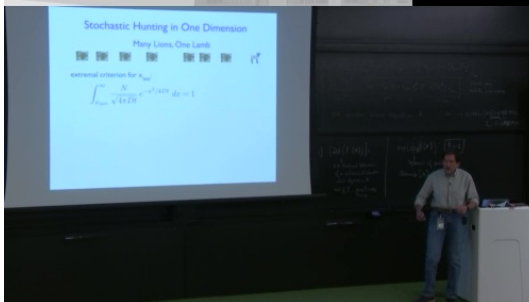
# Stochastic Hunting in One Dimension

Many Lions, One Lamb



extremal criterion for  $x_{\text{last}}$ :

$$\int_{x_{\text{last}}}^{\infty} \frac{N}{\sqrt{4\pi Dt}} e^{-x^2/4Dt} dx = 1$$



# Stochastic Hunting in One Dimension

## Many Lions, One Lamb



extremal criterion for  $x_{\text{last}}$ :

$$\int_{x_{\text{last}}}^{\infty} \frac{N}{\sqrt{4\pi Dt}} e^{-x^2/4Dt} dx = 1$$

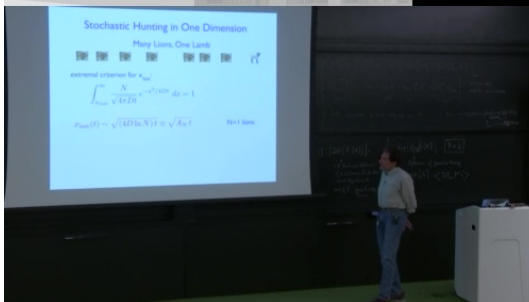
$$x_{\text{last}}(t) \sim \sqrt{(4D \ln N) t} \equiv \sqrt{A_N t}$$

$N \gg 1$  lions

$$x_{\text{last}}(t) \sim \sqrt{2D \ln(c_0^2 Dt) t}$$

$N = \infty$  lions

constant density for  $x < 0$   
only  $N \sim c_0 \sqrt{Dt}$  are “dangerous”



# Stochastic Hunting in One Dimension

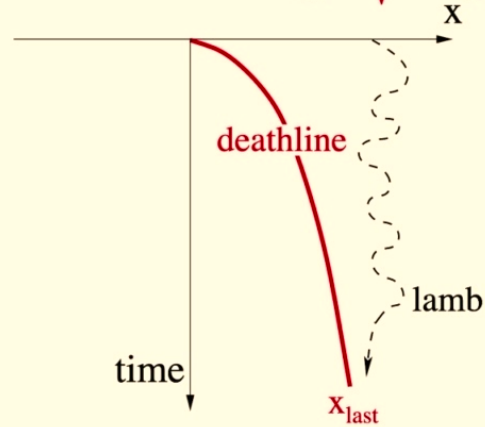
## Many Lions, One Lamb



Effective Problem:  
Deterministic Deadline

$$x_{\text{last}}(t) \sim \sqrt{(4D \ln N) t} \equiv \sqrt{A_N t}$$

$$x_{\text{last}}(t) \sim \sqrt{2D \ln(c_0^2 D t)} t \quad N = \infty$$



# Stochastic Hunting in One Dimension

## Many Lions, One Lamb

lamb probability distribution:

$$\frac{\partial p(x, t)}{\partial t} - \frac{x_{\text{last}}}{2t} \frac{\partial p(x, t)}{\partial x} = D \frac{\partial^2 p(x, t)}{\partial x^2} \quad (0 \leq x < \infty)$$

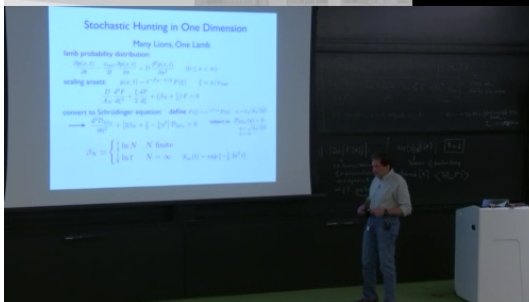
scaling ansatz:  $p(x, t) \sim t^{-\beta_N - 1/2} F(\xi) \quad \xi = x/x_{\text{last}}$

$$\frac{D}{A_N} \frac{d^2 F}{d\xi^2} + \frac{\xi}{2} \frac{dF}{d\xi} + \left(\beta_N + \frac{1}{2}\right) F = 0$$

convert to Schrödinger equation: define  $F(\xi) = e^{-\eta^2/4} \mathcal{D}(\eta) \quad \eta = \xi \sqrt{A_N/2D}$

$$\longrightarrow \frac{d^2 \mathcal{D}_{2\beta_N}}{d\eta^2} + \left[2\beta_N + \frac{1}{2} - \frac{1}{4}\eta^2\right] \mathcal{D}_{2\beta_N} = 0 \quad \text{subject to } \begin{cases} \mathcal{D}_{2\beta_N}(\eta) = 0 \\ \eta = \sqrt{A_N/2D} \\ \eta = \infty \end{cases}$$

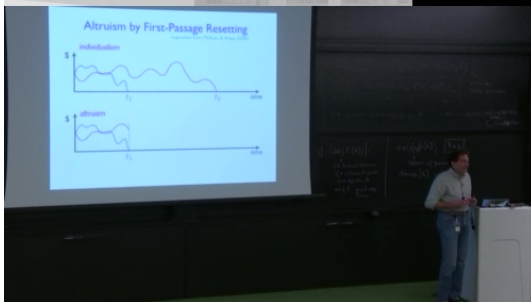
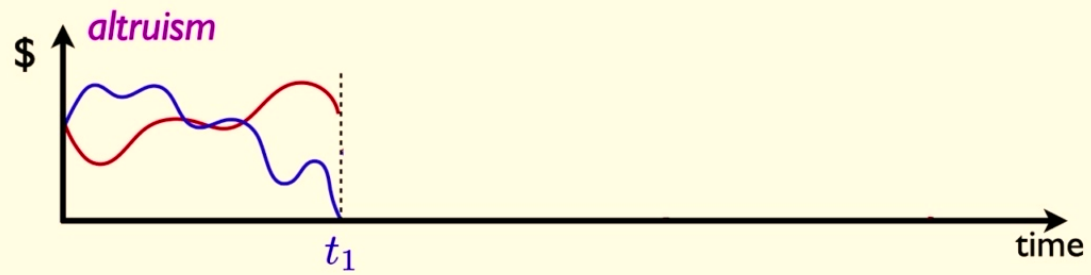
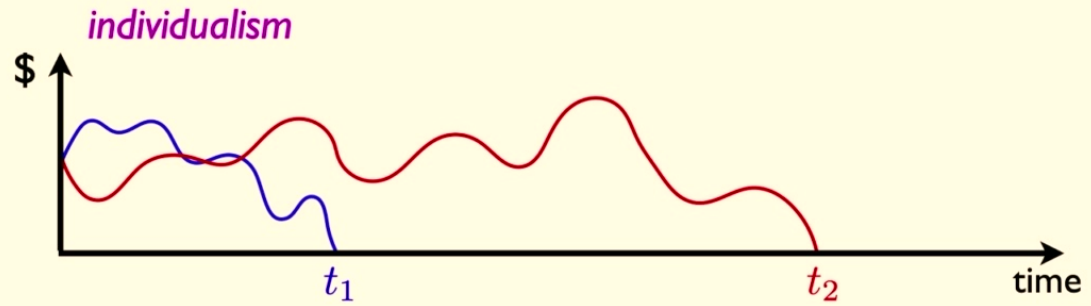
$$\beta_N \simeq \begin{cases} \frac{1}{4} \ln N & N \text{ finite} \\ \frac{1}{8} \ln t & N = \infty \end{cases} \quad S_\infty(t) \sim \exp\left(-\frac{1}{8} \ln^2 t\right)$$





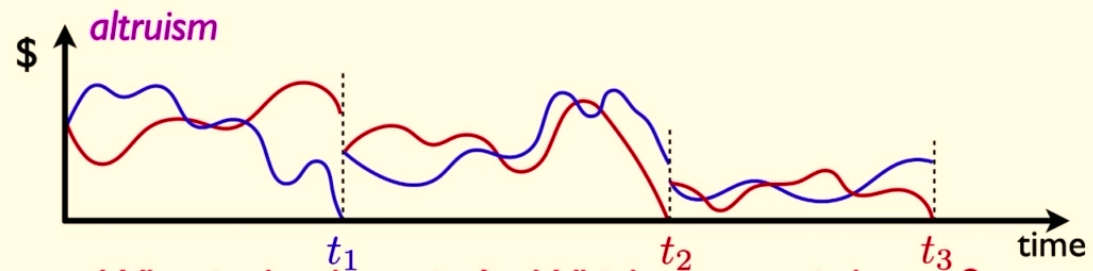
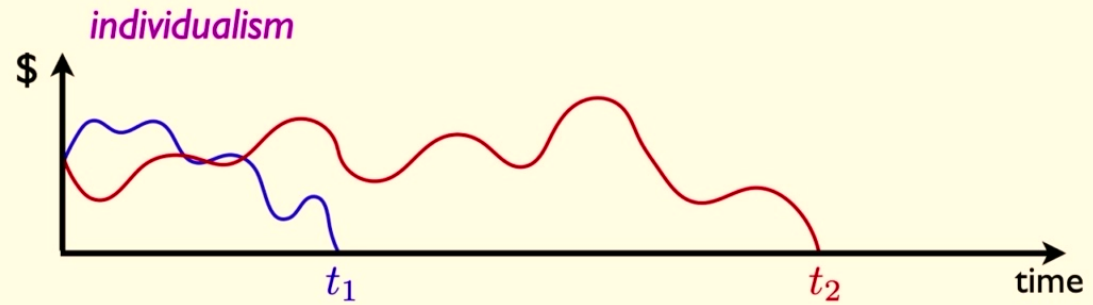
# Altruism by First-Passage Resetting

inspiration from McKean & Shepp (2006)

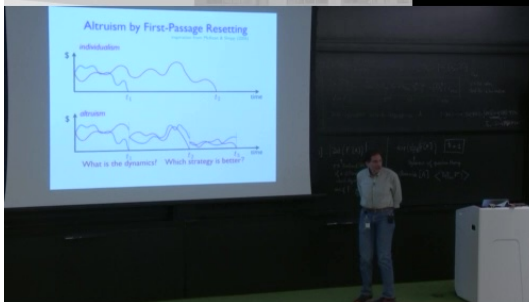


# Altruism by First-Passage Resetting

inspiration from McKean & Shepp (2006)



What is the dynamics? Which strategy is better?



# Dynamics of Individualism

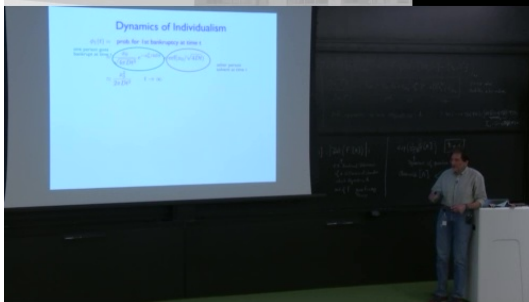
$\phi_1(t) =$  prob. for 1st bankruptcy at time  $t$

one person goes bankrupt at time  $t$

$$\frac{x_0}{\sqrt{4\pi Dt^3}} e^{-x_0^2/4Dt} \times \text{erf}(x_0/\sqrt{4Dt})$$

other person solvent at time  $t$

$$\simeq \frac{x_0^2}{2\pi Dt^2} \quad t \rightarrow \infty$$



# Dynamics of Individualism

$\phi_1(t) =$  prob. for 1st bankruptcy at time  $t$

one person goes bankrupt at time  $t$

$$= \frac{x_0}{\sqrt{4\pi Dt^3}} e^{-x_0^2/4Dt} \times \text{erf}(x_0/\sqrt{4Dt})$$

other person solvent at time  $t$

$$\simeq \frac{x_0^2}{2\pi Dt^2} \quad t \rightarrow \infty$$

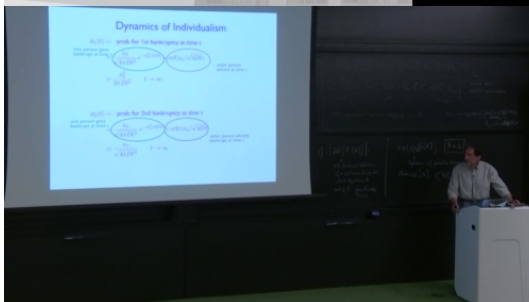
$\phi_2(t) =$  prob. for 2nd bankruptcy at time  $t$

one person goes bankrupt at time  $t$

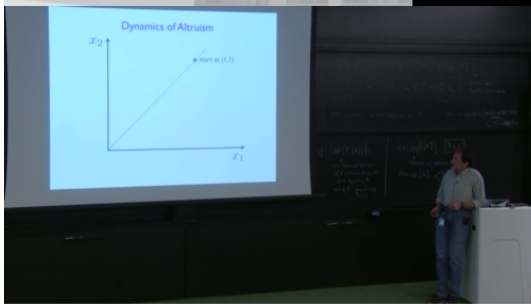
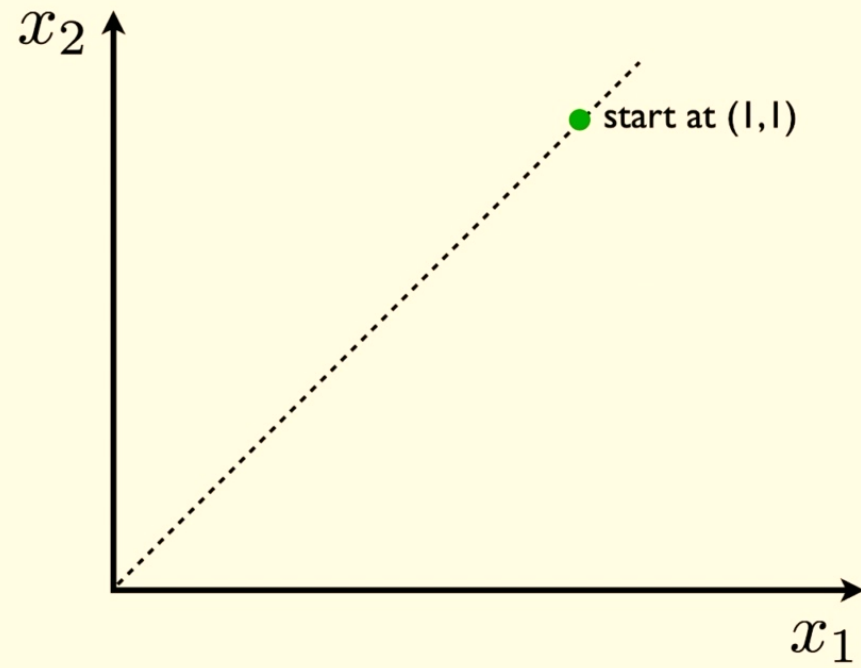
$$= \frac{x_0}{\sqrt{4\pi Dt^3}} e^{-x_0^2/4Dt} \times \text{erfc}(x_0/\sqrt{4Dt})$$

other person already bankrupt at time  $t$

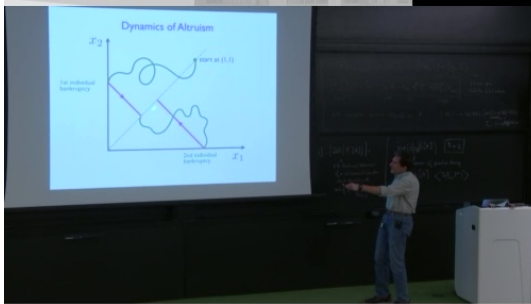
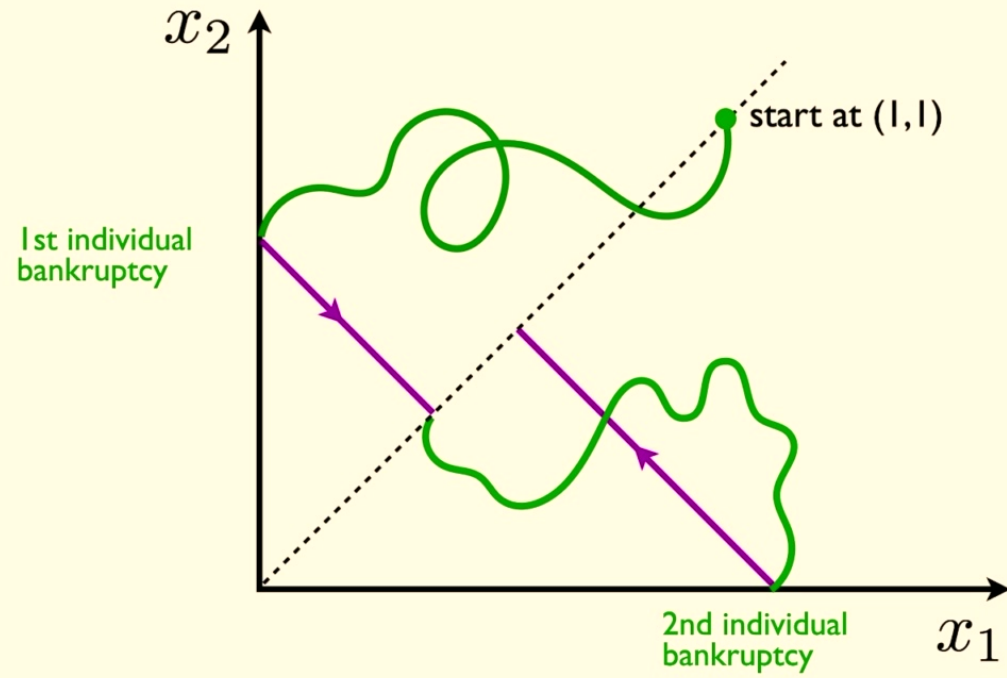
$$\simeq \frac{x_0}{\sqrt{4\pi Dt^3}} \quad t \rightarrow \infty$$



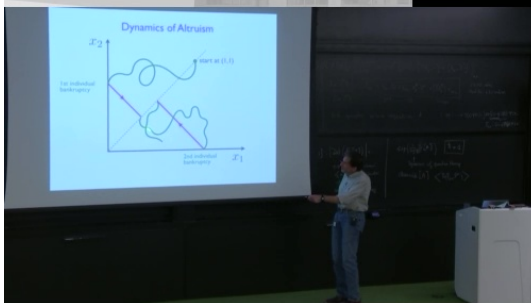
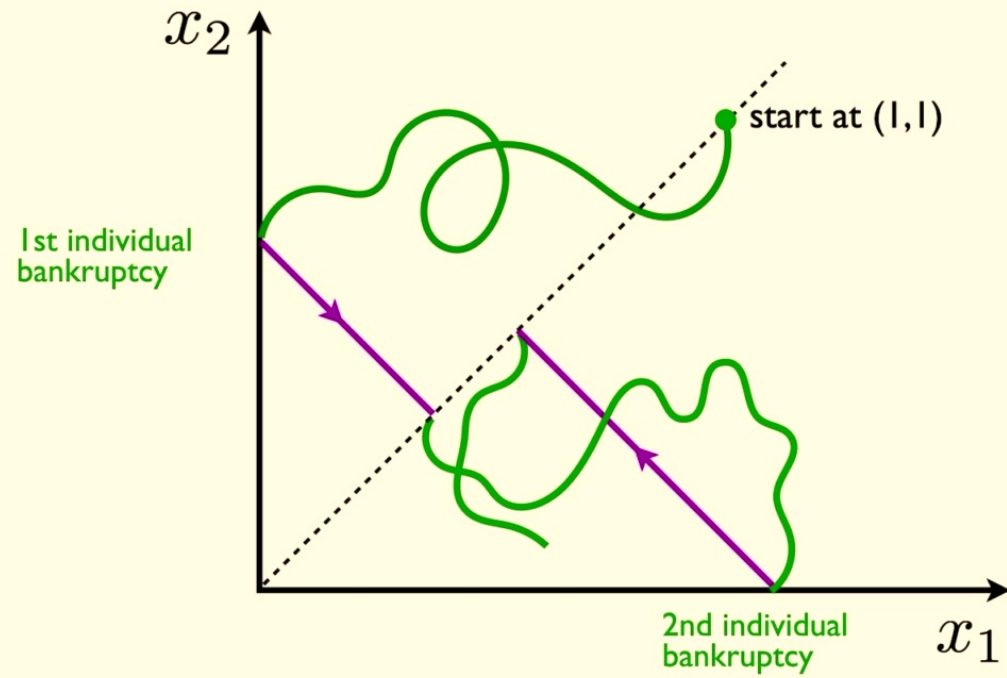
## Dynamics of Altruism



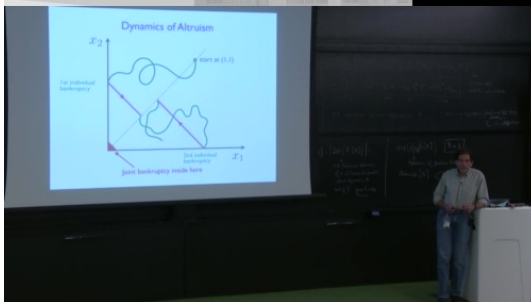
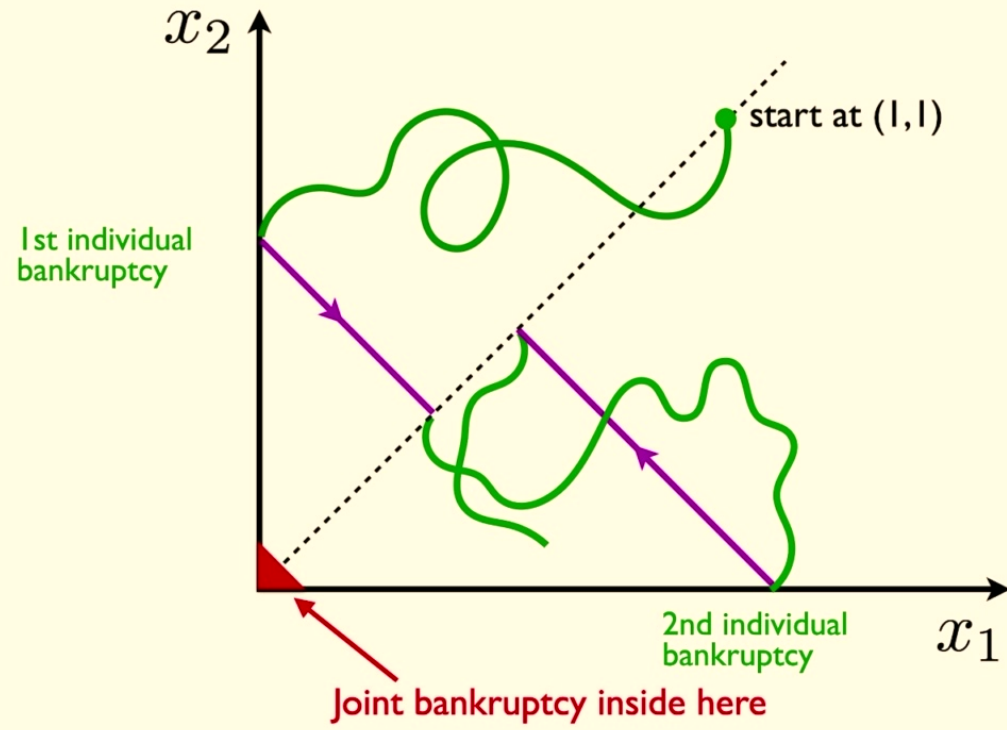
# Dynamics of Altruism



# Dynamics of Altruism



# Dynamics of Altruism





# Dynamics of Altruism

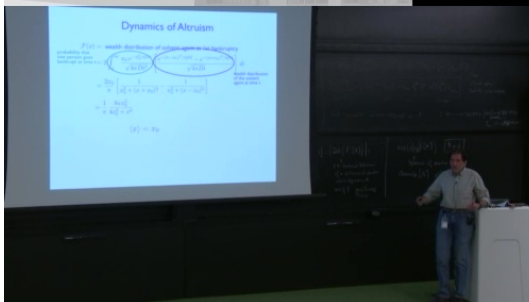
$\mathcal{F}(x)$  = wealth distribution of solvent agent at 1st bankruptcy  
 probability that one person goes bankrupt at time  $t$  =  $2 \int_0^\infty \frac{x_0 e^{-x_0^2/4Dt}}{\sqrt{4\pi Dt^3}} \left[ \frac{e^{-(x-x_0)^2/4Dt} - e^{-(x+x_0)^2/4Dt}}{\sqrt{4\pi Dt}} \right] dt$

wealth distribution of the solvent agent at time  $t$

$$= \frac{2x_0}{\pi} \left[ \frac{1}{x_0^2 + (x+x_0)^2} - \frac{1}{x_0^2 + (x-x_0)^2} \right]$$

$$= \frac{1}{\pi} \frac{8xx_0^2}{4x_0^4 + x^4}$$

$$\langle x \rangle = x_0$$



# Dynamics of Altruism

$\mathcal{F}(x)$  = wealth distribution of solvent agent at 1st bankruptcy  
 probability that one person goes bankrupt at time  $t$  =  $2 \int_0^\infty \frac{x_0 e^{-x_0^2/4Dt}}{\sqrt{4\pi Dt^3}} \left[ \frac{e^{-(x-x_0)^2/4Dt} - e^{-(x+x_0)^2/4Dt}}{\sqrt{4\pi Dt}} \right] dt$

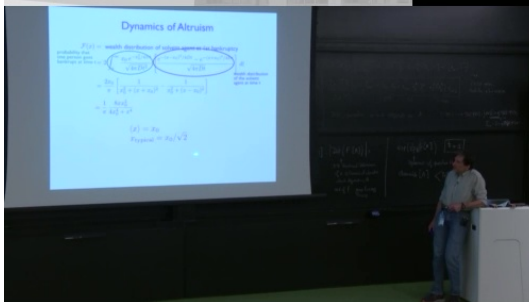
wealth distribution of the solvent agent at time  $t$

$$= \frac{2x_0}{\pi} \left[ \frac{1}{x_0^2 + (x+x_0)^2} - \frac{1}{x_0^2 + (x-x_0)^2} \right]$$

$$= \frac{1}{\pi} \frac{8xx_0^2}{4x_0^4 + x^4}$$

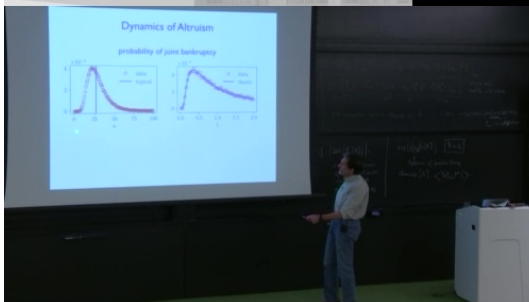
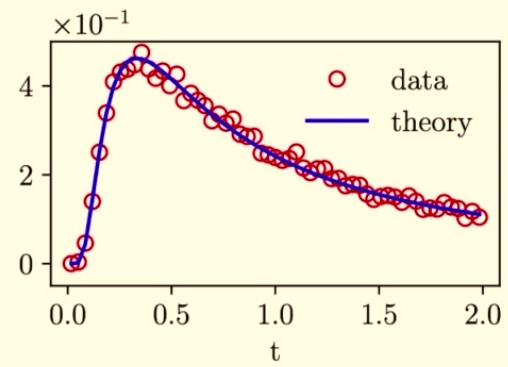
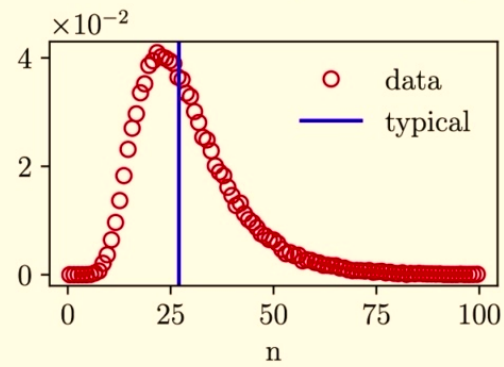
$$\langle x \rangle = x_0$$

$$x_{\text{typical}} = x_0/\sqrt{2}$$

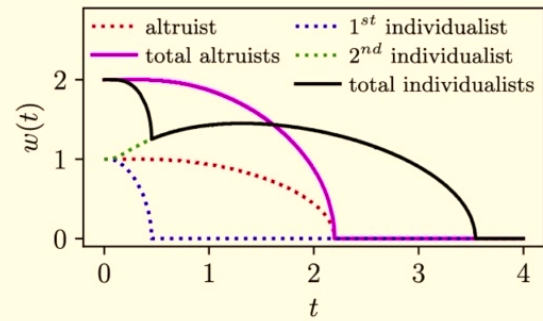


# Dynamics of Altruism

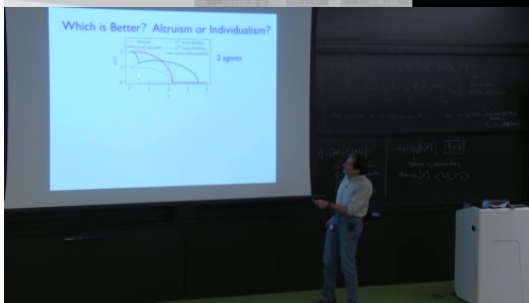
probability of joint bankruptcy



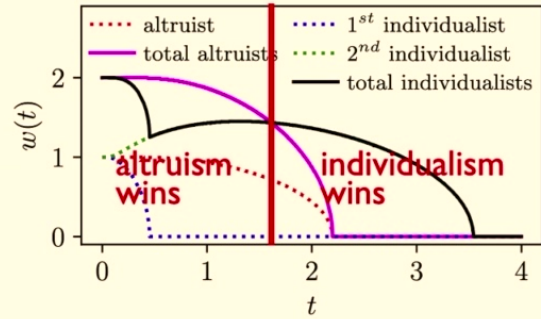
# Which is Better? Altruism or Individualism?



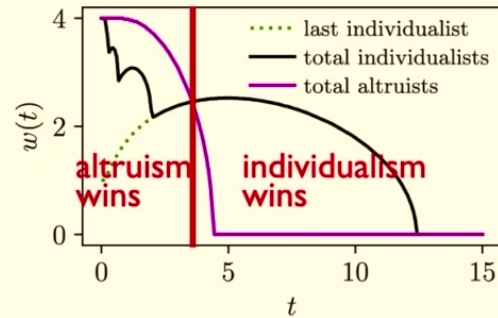
2 agents



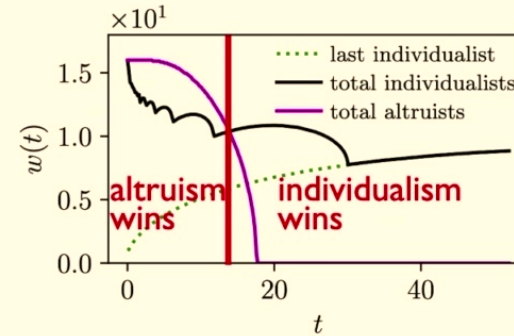
# Which is Better? Altruism or Individualism?



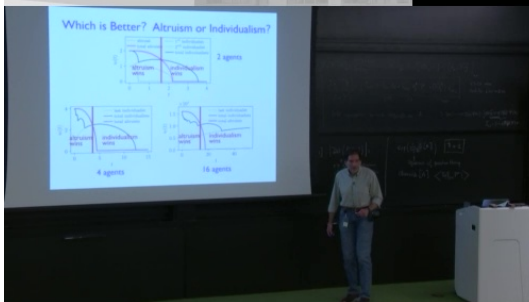
2 agents



4 agents

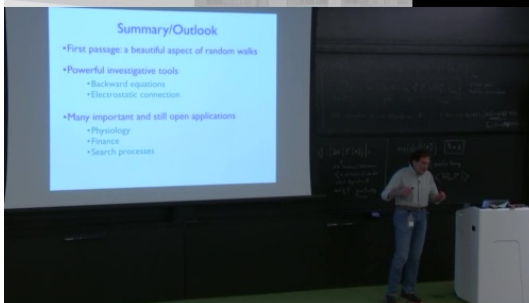


16 agents



# Summary/Outlook

- First passage: a beautiful aspect of random walks
- Powerful investigative tools
  - Backward equations
  - Electrostatic connection
- Many important and still open applications
  - Physiology
  - Finance
  - Search processes



Thanks for listening!

